# → TCS RIO 125 Internship project

## Automated Sentiment Analysis of textual comments and feedback

#### **Refrance Links from tcs**

- 1. <a href="https://www.intel.com/content/www/us/en/developer/topic-technology/artificial-intelligence/training/courses.html">https://www.intel.com/content/www/us/en/developer/topic-technology/artificial-intelligence/training/courses.html</a>
- 2. <a href="https://python-course.eu/machine-learning/">https://python-course.eu/machine-learning/</a>
- 3. https://scikit-learn.org/stable/tutorial/index.html
- 4. <a href="https://pandas.pydata.org/pandas-docs/version/0.15/tutorials.html">https://pandas.pydata.org/pandas-docs/version/0.15/tutorials.html</a>
- 5. <a href="https://www.tensorflow.org/tutorials">https://www.tensorflow.org/tutorials</a>
- 6. https://www.tensorflow.org/tutorials/keras/classification
- 7. <a href="https://www.tensorflow.org/tutorials/images/cnn">https://www.tensorflow.org/tutorials/images/cnn</a>
- 8. https://www.tensorflow.org/text/tutorials/text\_classification\_rnn
- 9. https://www.intel.com/content/www/us/en/developer/learn/course-deep-learning.html
- 10. https://www.intel.com/content/www/us/en/developer/learn/course-machine-learning.html

#### ▼ Importing libraries

```
1 import numpy as np
2 import pandas as pd
3 print('done')

done
```

#### Importing dataset

1 dataset.shape

```
1 from google.colab import drive
2 drive.mount('/content/drive')
   Mounted at /content/drive

1 %cd /content/drive/MyDrive/Msc AI /TCS iON Rio -125 internship
2 !ls
   /content/drive/MyDrive/Msc AI /TCS iON Rio -125 internship
   a1_RestaurantReviews_HistoricDump.csv 'Sentiment Analysis P1.ipynb'
   a2_RestaurantReviews_FreshDump.csv

1 dataset = pd.read_csv('/content/drive/MyDrive/Msc AI /TCS iON Rio -125 int
```

```
(900, 2)
```

### 1 dataset.head(10)

```
1
                                              Review Liked
0
                             Wow... Loved this place.
                                    Crust is not good.
                                                            0
             Not tasty and the texture was just nasty.
                                                            0
     Stopped by during the late May bank holiday of...
   The selection on the menu was great and so wer...
    Now I am getting angry and I want my damn pho.
                                                            0
                  Honeslty it didn't taste THAT fresh.)
     The potatoes were like rubber and you could te...
                             The fries were great too.
                                       A great touch.
```

### 1 dataset['Review'].head()

```
Wow... Loved this place.
Crust is not good.
Not tasty and the texture was just nasty.
Stopped by during the late May bank holiday of...
The selection on the menu was great and so wer...
Name: Review, dtype: object
```

### 1 dataset.shape

(900, 2)

#### ▼ Data Preprocessing

1 corpus=[]

```
1 import re
2 import nltk
3
4 nltk.download('stopwords')
5
6 from nltk.corpus import stopwords
7 from nltk.stem.porter import PorterStemmer
8 ps = PorterStemmer()
9
10 all_stopwords = stopwords.words('english')
11 all_stopwords.remove('not')

[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

```
2
3 for i in range(0, 900):
4    review = re.sub('[^a-zA-Z]', ' ', dataset['Review'][i])
5    review = review.lower()
6    review = review.split()
7    review = [ps.stem(word) for word in review if not word in set(all_stopwc review = ' '.join(review)
9    corpus.append(review)
```

#### 1 corpus

```
['wow love place',
 'crust not good',
 'not tasti textur nasti',
 'stop late may bank holiday rick steve recommend love',
 'select menu great price',
 'get angri want damn pho',
 'honeslti tast fresh',
 'potato like rubber could tell made ahead time kept warmer',
 'fri great',
 'great touch',
 'servic prompt',
 'would not go back',
 'cashier care ever say still end wayyy overpr',
 'tri cape cod ravoli chicken cranberri mmmm',
 'disgust pretti sure human hair',
 'shock sign indic cash',
 'highli recommend',
 'waitress littl slow servic',
 'place not worth time let alon vega',
 'not like',
 'burritto blah',
 'food amaz',
 'servic also cute',
 'could care less interior beauti',
 'perform',
 'right red velvet cake ohhh stuff good',
 'name',
 'hole wall great mexican street taco friendli staff',
 'took hour get food tabl restaur food luke warm sever run around like total overwhelm',
 'worst salmon sashimi',
 'also combo like burger fri beer decent deal',
 'like final blow',
 'found place accid could not happier',
 'seem like good quick place grab bite familiar pub food favor look elsewher',
 'overal like place lot',
 'redeem qualiti restaur inexpens',
 'ampl portion good price',
 'poor servic waiter made feel like stupid everi time came tabl',
 'first visit hiro delight',
 'servic suck',
 'shrimp tender moist',
 'not deal good enough would drag establish',
 'hard judg whether side good gross melt styrofoam want eat fear get sick',
 'posit note server attent provid great servic',
 'frozen puck disgust worst peopl behind regist',
 'thing like prime rib dessert section',
 'bad food damn gener',
 'burger good beef cook right',
 'want sandwich go firehous',
 'side greek salad greek dress tasti pita hummu refresh',
 'order duck rare pink tender insid nice char outsid',
 'came run us realiz husband left sunglass tabl',
 'chow mein good',
 'horribl attitud toward custom talk one custom enjoy food',
```

```
'portion huge',
'love friendli server great food wonder imagin menu',
'heart attack grill downtown vega absolut flat line excus restaur',
```

#### ▼ Data transformation

1 y

```
array([1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1,
       1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1,
      0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1,
      1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1,
      1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
      0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1,
      1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0,
      1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0,
      0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0,
      1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1,
       0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1,
       0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1,
      1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1, 1,
       0, 0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0,
      1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 0,
      0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0,
      1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1,
       0, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1,
       0, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1,
      1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1,
       0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1,
      1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0,
      1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0,
       0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1,
      1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1,
      1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 0,
       0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 0,
       0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1,
      0, 0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1,
       0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1,
      1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0,
      1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1,
      1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1,
      1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1,
       0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 1,
      0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0,
       0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0,
      1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0,
```

```
0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1])
  1 type(y)
    numpy.ndarray
  1 # Saving BoW dictionary to later use in prediction
  2 import pickle
  3 bow_path = 'c1_BoW_Sentiment_Model.pkl'
  4 pickle.dump(cv, open(bow_path, "wb"))
Dividing dataset into training and test set
 [ ] L, 5 cells hidden
Model fitting

→ SVM
  1 # SVM model
  2 #Import svm model
  3 from sklearn import svm
  4 #Create a svm Classifier
  5 clf = svm.SVC(kernel='linear') # Linear Kernel
  6 #Train the model using the training sets
  7 clf.fit(X train, y train)
  8 #Predict the response for test dataset
  9 Y pred = clf.predict(X_test)
  1 print(Y_pred)
    1 Y_pred.shape
    (180,)
  1 import joblib
  2 joblib.dump(clf, 'Clf Sentiment Model SVM')
    ['Clf_Sentiment_Model_SVM']
  1 type(Y_pred)
```

1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1,

0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0,

numpy.ndarray

```
1 df = pd.DataFrame(Y_pred, columns = ['Polarity'])
2 df.head(6)
```

	Polarity
0	1
1	1
2	0
3	1
4	0
5	0

```
1 df.shape
(180, 1)
```

```
1 df.info()
```

```
1 df.count()
```

Polarity 180

```
1 # Exporting SVM to later use in prediction
2 import joblib
3 joblib.dump(clf, 'Clf_Sentiment_Model')
```

```
['Clf_Sentiment_Model']
```

## Naive Bays

```
1 # importing Naive Bays model from scikit-learn
2 from sklearn.naive_bayes import GaussianNB
3 classifier = GaussianNB()
4 classifier.fit(X_train, y_train)
5 y_pred = classifier.predict(X_test)

1 X_train.shape
(720, 1420)
```

```
1 y train.shape
      (720,)
   1 len(y_pred)
      180
   1 # Exporting NB Classifier to later use in prediction
   2 import joblib
   3 joblib.dump(classifier, 'c2 Classifier Sentiment Model')
      ['c2_Classifier_Sentiment_Model']
Model performance
   1 type(y_test)
      numpy.ndarray
   1 #Import scikit-learn metrics module for accuracy calculation
   2 from sklearn.metrics import confusion matrix, accuracy score
   3 # Model Accuracy: how often is the classifier correct?
   4 acc =accuracy score(y test, Y pred)
   5 cm = confusion matrix(y test,Y pred)
    6 print('confusion matrix:\n', cm )
   7 print('Accuracy % :',acc*100)
      confusion matrix:
       [[58 20]
       [20 82]]
      Accuracy %: 77.77777777779
   1 y test
      1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1,
            0, 0, 1, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1,
            0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0,
            1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 0, 0,
            1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 1,
            1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 0, 1, 1, 1,
            1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1,
            0, 0, 0, 0])
   1 y pred
      array([0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0,
            0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 1,
            0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1,
            0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0,
            1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0,
            0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1,
            1, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1,
```

```
0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0])
```

```
1 from sklearn.metrics import confusion_matrix, accuracy_score
2 cm = confusion_matrix(y_test, y_pred)
3 # print(cm)
4 ACc = accuracy_score(y_test, y_pred)
5 print('Accuracy:',ACc*100)
6 print('Confusion Matx:\n',cm)
```

Accuracy: 73.88888888888889 Confusion Matx: [[67 11] [36 66]]

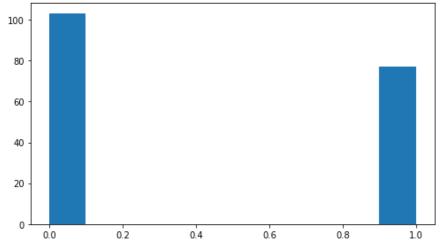
#### "Model accuracy is better with SVM than NAIVE BAYS algorithm"

**Accuracy Scores** 

- 1. SVM =77.7 ~ 78%
- 2. Naive Bays = 73.8 ~ 74

#### 1 plt.hist(y\_pred)

```
(array([103., 0., 0., 0., 0., 0., 0., 0., 0., 77.]), array([0., 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1.]), <a list of 10 Patch objects>)
```



## → Cross Validation

1 %cd /content/drive/MyDrive/Msc AI /TCS iON Rio -125 internship 2 !ls

```
/content/drive/MyDrive/Msc AI /TCS iON Rio -125 internship a1_RestaurantReviews_HistoricDump.csv 'Sentiment Analysis P1.ipynb' a2_RestaurantReviews_FreshDump.csv
```

1 q = '/content/drive/MyDrive/Msc AI /TCS iON Rio -125 internship/a2\_Restaur

```
1 data = pd.read_csv(q)
2 data.head()
                                      Review
    0
                    Spend your money elsewhere.
    1 Their regular toasted bread was equally satisf...
    2 The Buffet at Bellagio was far from what I ant...
                 And the drinks are WEAK, people!
                        -My order was not correct
1 data.count
   <bound method DataFrame.count of</pre>
                                                                                 Review
                            Spend your money elsewhere.
       Their regular toasted bread was equally satisf...
      The Buffet at Bellagio was far from what I ant...
                       And the drinks are WEAK, people!
                             -My order was not correct.
      I think food should have flavor and texture an...
   95
   96
                               Appetite instantly gone.
   97 Overall I was not impressed and would not go b...
   98 The whole experience was underwhelming, and I ...
   99 Then, as if I hadn't wasted enough of my life ...
   [100 rows x 1 columns]>
1 data.shape
   (100, 1)
1 data.describe
   <bound method NDFrame.describe of</pre>
                                                                                  Review
                            Spend your money elsewhere.
      Their regular toasted bread was equally satisf...
   1
      The Buffet at Bellagio was far from what I ant...
                       And the drinks are WEAK, people!
                             -My order was not correct.
   95 I think food should have flavor and texture an...
                               Appetite instantly gone.
   97 Overall I was not impressed and would not go b...
   98 The whole experience was underwhelming, and I ...
   99 Then, as if I hadn't wasted enough of my life ...
   [100 rows x 1 columns]>
1 import re
2 import nltk
3
4 nltk.download('stopwords')
6 from nltk.corpus import stopwords
7 from nltk.stem.porter import PorterStemmer
8 ps = PorterStemmer()
```

```
10 all_stopwords = stopwords.words('english')
11 all stonwords.remove('not')
   [nltk_data] Downloading package stopwords to /root/nltk_data...
   [nltk_data] Unzipping corpora/stopwords.zip.
1 corpus=[]
2
3 for i in range(0, 100):
   review = re.sub('[^a-zA-Z]', ' ', data['Review'][i])
   review = review.lower()
   review = review.split()
   review = [ps.stem(word) for word in review if not word in set(all stopwo
   review = ' '.join(review)
   corpus.append(review)
1 # Loading BoW dictionary
2 from sklearn.feature extraction.text import CountVectorizer
3 import pickle
4 cvFile='c1 BoW Sentiment Model.pkl'
5 # cv = CountVectorizer(decode_error="replace", vocabulary=pickle.load(open
6 cv = pickle.load(open(cvFile, "rb"))
1 X_fresh = cv.transform(corpus).toarray()
2 X fresh.shape
   (100, 1420)
1 import joblib
2 classifier = joblib.load('c2 Classifier Sentiment Model')
3 clf = joblib.load('Clf_Sentiment_Model_SVM')
1 yf pred = classifier.predict(X fresh)
2 print(yf_pred)
   1 yf2 pred = clf.predict(X fresh)
2 print(yf2 pred)
   0 1 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 1 0 0 0 1 0 1 0 1
1 data['predicted label'] = yf pred.tolist()
 2 data.head(10)
```

```
1 data['predicted_label'] = yf2_pred.tolist()
2 data hoad(10)

1 data.to_csv('file1.csv',index= False)

1 len(y_test)
180

1 len(yf_pred)
100

1 len(yf2_pred)
100
```

## ▼ Result Visulization

```
1 import matplotlib.pyplot as plt

1 df = pd.read_csv('file1.csv')

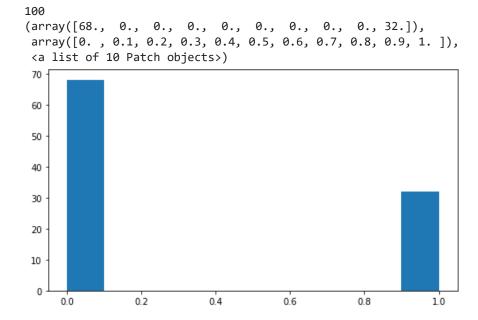
1 df.head()

1 plt.rcParams["figure.figsize"] = [7,4]
2 plt.rcParams["figure.autolayout"] = True
3 plt.title("Predicted values on unseen data")
4 # plt.hist(df['predicted_label'].head())

1 print(len(y_pred))
2 plt.title("y_pred viz",c='r')
3 plt.hist(y_pred)
```

```
1 print(len(yf_pred))
2 plt.hist(yf_pred)
```

```
print(len(yf2_pred))
plt.hist(yf2_pred)
```



We can conclud from the above plots that most of the feedbacks are negative feedbacks

# End of The Project

• ×

✓ 0s completed at 11:42 PM